SAIC Analysis of Data Acquired at Camp Butner, NC

Dean Keiswetter



Partners in Environmental Technology Technical Symposium and Workshop; Washington, DC, Nov 30 – Dec 2, 2010

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14. ABSTRACT

The Large Scale Classification Project at Camp Butner provides an excellent opportunity to compare and contrast classification performances for static and reconnaissance EMI data and for a variety of analysis approaches. SAIC analyzed EM61 data acquired in reconnaissance mode as well as Metal Mapper and TEMTADS data acquired while stationary. Our analysis included single- and multi-source solvers. Our classification utilizes a decision tree targeting the intrinsic polarizabilities. The decision tree incorporates uncertainty in unanticipated targets-ofinterest and has hasn?t changed dramatically since being developed using data acquired at Aberdeen Proving Ground, Camp Sibert, and Camp San Luis Obispo. We also experimented in the number of training labels (starting with no on-site labels) used to fine tune the classifier. Finally, we utilized two different analysis environments; Oasis montaj and IDL. Two commercial firms, NAEVA and Parsons, also utilized the UX-Analyze module in Oasis montaj to classify Metal Mapper stationary data. During our presentation, we will discuss performances of the various combinations and present lessons learned.

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SAIC DATA ANALYSIS OF DATA ACQUIRED AT CAMP BUTNER

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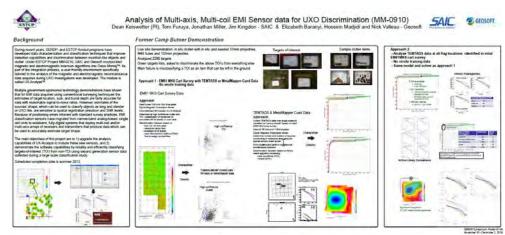


Outline

Background

Datasets analyzed
Analysis environment
Inversion schemes
Classification approach
EM61 data as pre-screen
Classification performance
Failure Analysis

Stop by Poster #61 for more details





Project Team & Sponsor

Science Applications International Corporation (SAIC)

Dean Keiswetter program manager

Tom Furuya data analyst

Jim Kingdon data analyst & analysis algorithms

Nagi Khadr data analyst

Jonathan Miller analysis algorithms

Bruce Barrow failure analysis

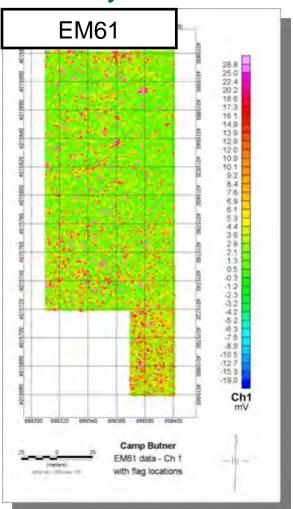
Tom Bell technical advisor

Supported by ESTCP Project's MM-0910 & MM-0134

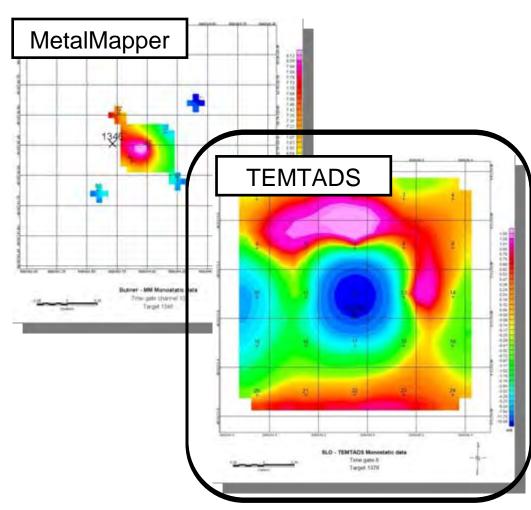


Sensor Data

Dynamic



Cued





Analysis Environment

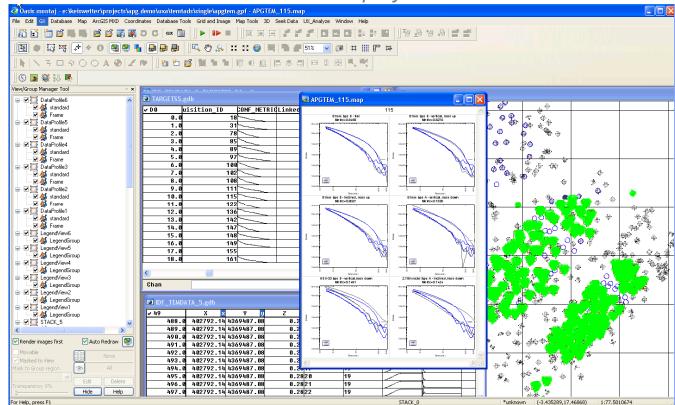


Oasis montaj

- High performance database
- Advanced data processing
- Dynamic linking (maps, data, profiles, etc.)
- Professional map production
- Audit trail

UX-Analyze

Stop by Poster #60 for more details





Single target solvers

Standard dipole model

 Location (X,Y,Z), orientation (Ψ, Θ, Φ), & intrinsic polarizabilities

Utilized two single source, but multi-stage solvers – each designed to avoid local minima

- ♦ Generally produce the same answers
- Subtle difference in recovered polarizabilities are sometimes observed
- Excellent data for establishing best practices

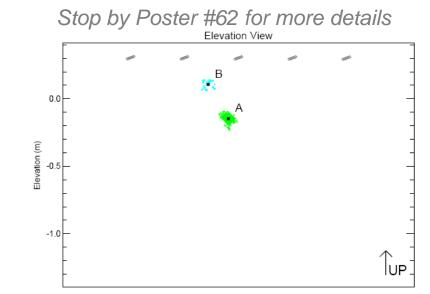


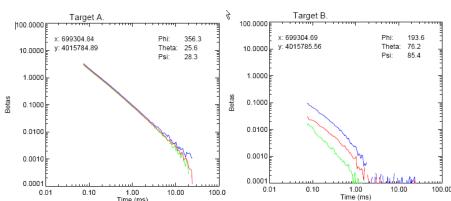
Multi-source Solver

Multi-source solver for handling multiple objects within the sensors' field of view (MM-1662)

- ♦ Seed the area with sources
- Predict signals with forward model
- Find a linear combination that best match observed signal using sparse solution solver
- Add new seeds
- ♦ Iterate

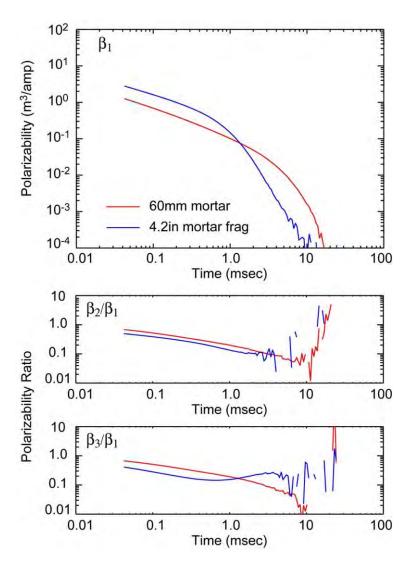
Perform multi-dipole inversion on derived target locations







Classification Approach

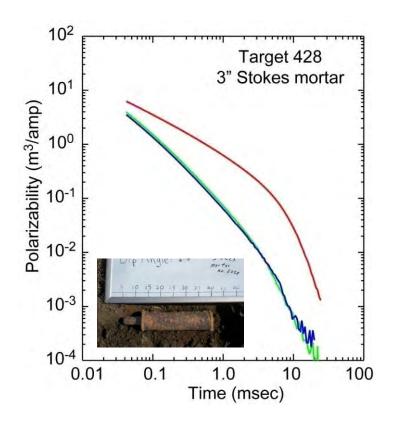


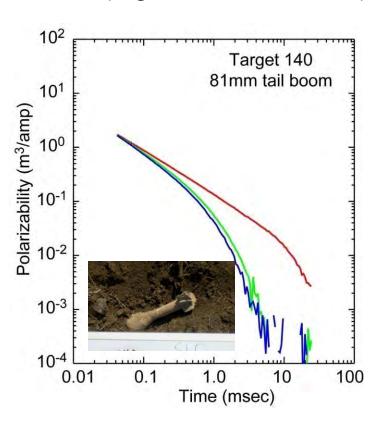
- Compare unconstrained polarizabilities for the target under investigation to a signature library
- "Library match" metric
 - 1. Primary polarizability (β_1)
 - 2. Ratio secondary to primary (β_2/β_1)
 - 3. Ratio tertiary to primary (β_3/β_1)
- Decision boundary chosen to accommodate training data



Axial Symmetry

- Targets with axially symmetric response that do not match expected munitions included in "can't decide"
 - ♦ Hedge against unexpected munitions (e.g. 3" Stokes mortar)

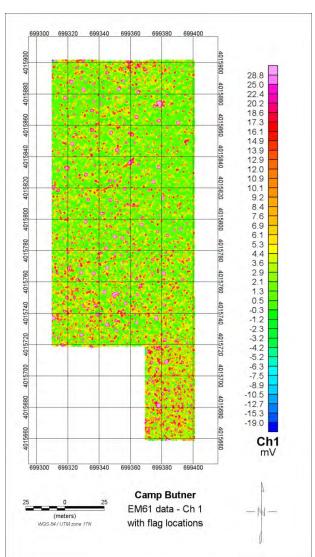






EM61 as pre-screener

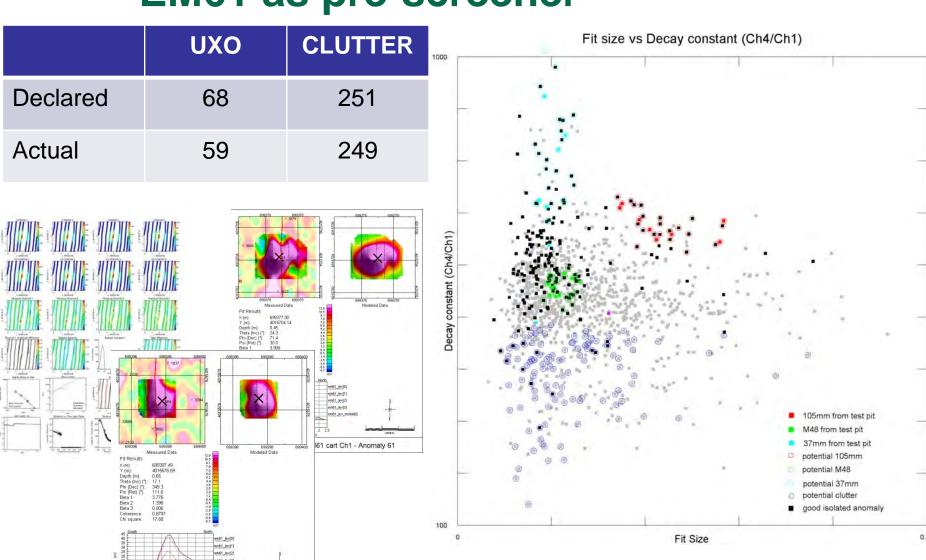
- Lower coil only, four gates
- Unconstrained 3-polarization
- Identified high confidence
 - UXO → dig
 - Clutter → leave
- All others request cued data
- Classification based on
 - Size (∑β 1st time gate)
 - Measured decay
 - Screen on fit quality
 - Generalized likelihood ratio test to assign probabilities





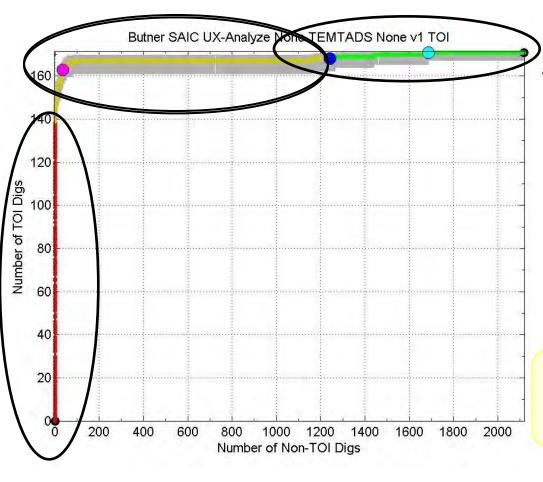
EM61 as pre-screener

Camp Butner - EM61 cart Ch1 - Anomaly 404





Classification Performance



2,290 anomalies

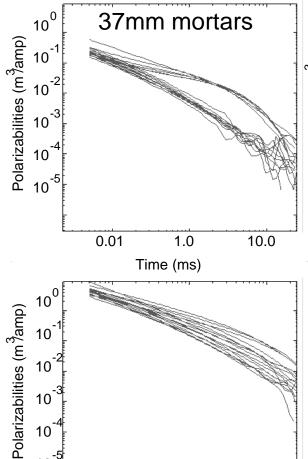
- 0 training
- 0 can't analyze
- 1,021 classified
 - 139/142 munitions correctly classified (97.9%)
 - 877/879 clutter correctly classified (99.8%)

1,269 can't decide

29 UXO, 1,240 clutter



Signature Variability



105mm projectile

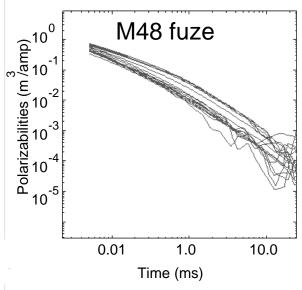
Time (ms)

10.0

0.01

10

10

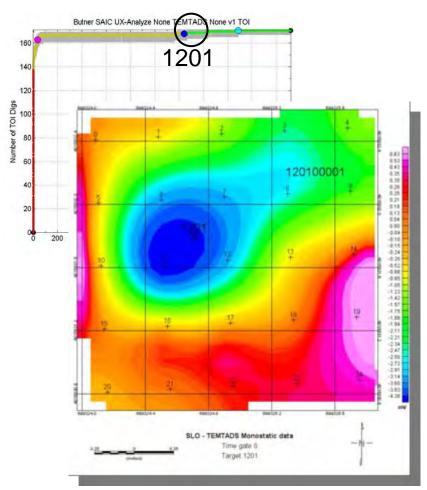


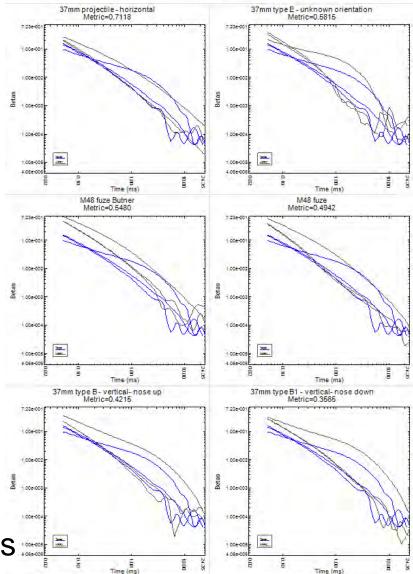
- Munitions in each class (37mm, M48, 105mm) are not identical
- Response curves can vary due to target condition: different model, fuze & tail boom present/absent, etc.

Correctly Classified as UXO				
37mm	118			
M48 fuze	24			
105mm	26			



Misclassified Munitions (1 of 3)



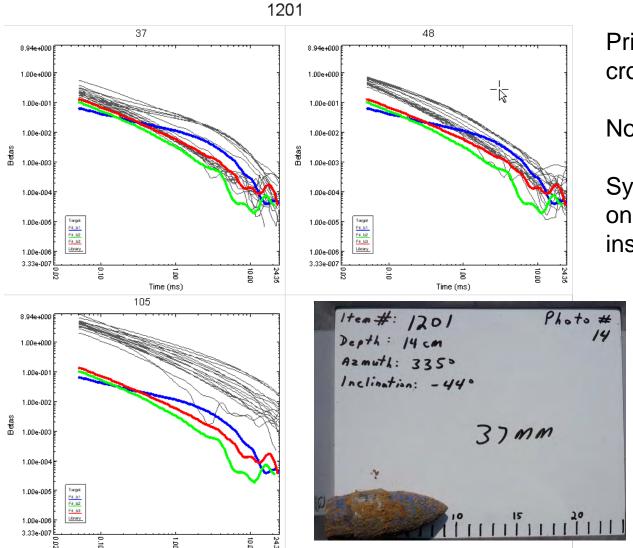


Signature Comparison:

ID 1201 versus library signatures



Misclassified Munitions (1 of 3)



Time (ms)

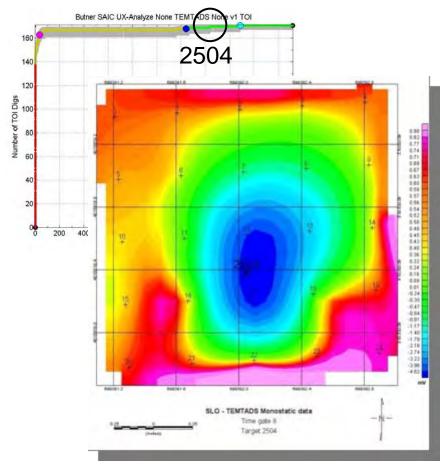
Principal polarizability crosses the other two

Not in our library

Symmetry metric based on polarizations 1 & 2 instead of 2 & 3

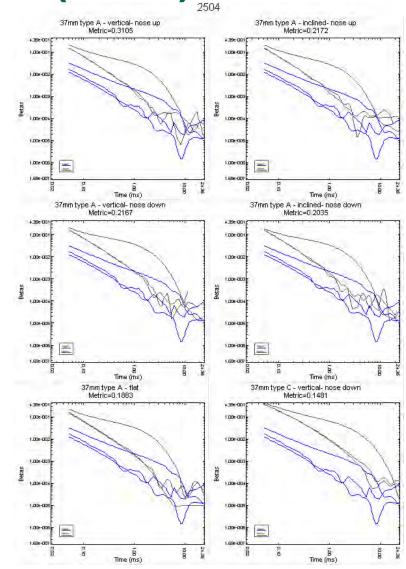


Misclassified Munitions (2 of 3)



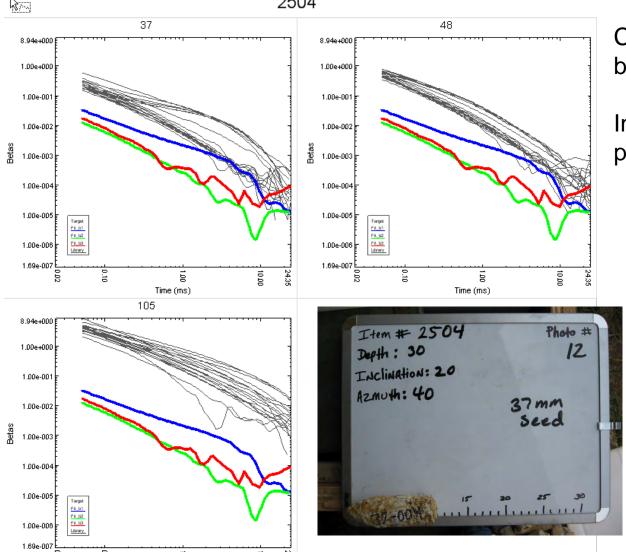
Signature Comparison:

ID 2504 versus library signatures





Misclassified Munitions (2 of 3)



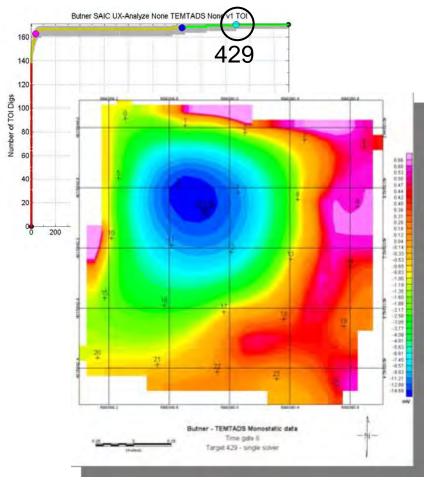
Time (ms)

Classified as clutter based on size

Inverted depth and polarizabilities too small

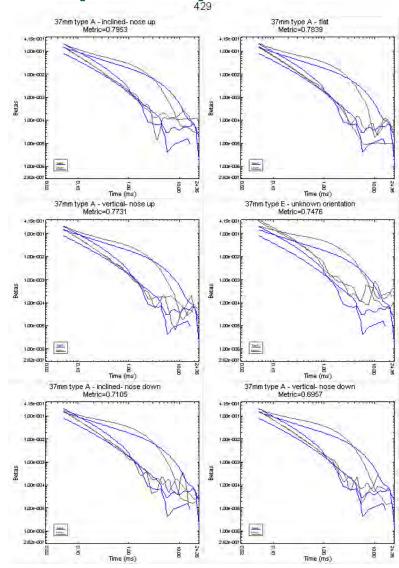


Misclassified Munitions (3 of 3)



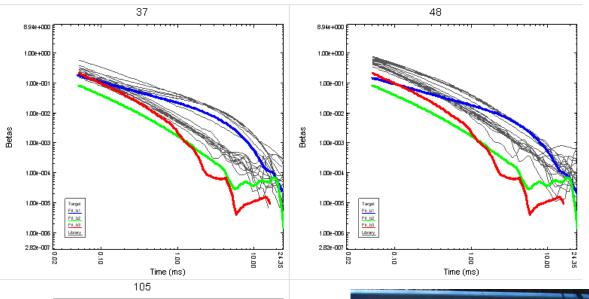
Signature Comparison:

ID 429 versus library signatures



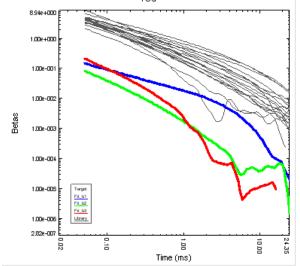


Misclassified Munitions (3 of 3)



Decision metric of 0.80, just below our threshold of 0.81.

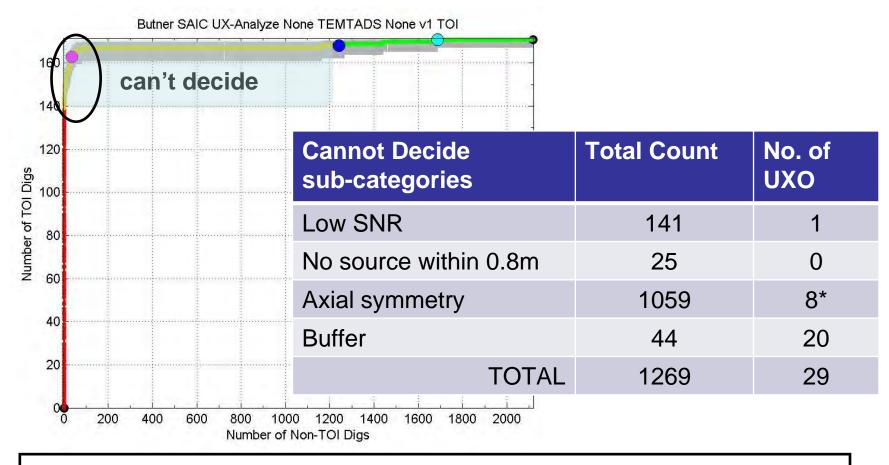
Decent signal strength put it in the high confidence clutter category







"Can't Decide" Category



*modifying our UXO/clutter threshold and not hedging for unexpected munitions types (viz., axial symmetry) would have reduced the unnecessary digs by 951

)



Summary/Conclusions

Our attempt to conservatively pre-screen using EM61 data (inverted size & measured decay) resulted in two false negatives

Classification based on intrinsic polarizabilities is effective

The vast majority of UXO were readily classified

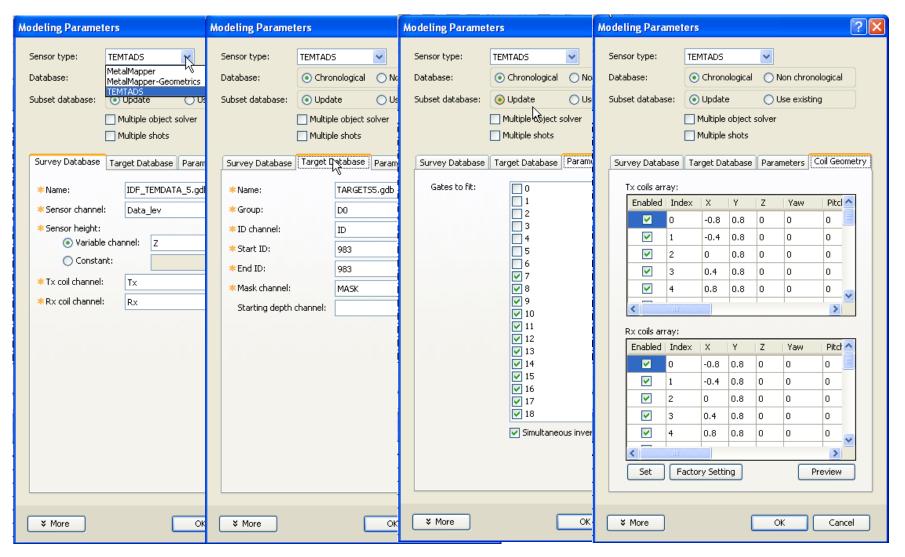
♦ 37mm showed the most variability and were the most difficult for us

Areas for classification performance improvement

- ♦ Low SNR targets Longer stacks, more robust classifier
- Multiple targets Adaptive array positioning, improved multi-target solvers
- ♦ Misclassified munitions Consolidate and adopt program-wide best practices for recognizing and dealing with outliers



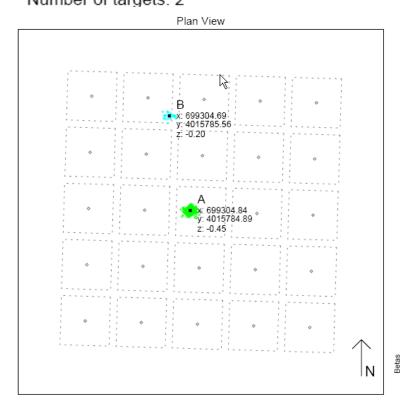
Analysis Interface





Solver Documentation (*.pdf)

NRL TEM array Number of targets: 2



Archive Documentation for each anomaly processed

